

# MTAEA – Matrix Algebra

Scott McCracken

School of Economics,  
Australian National University

January 19, 2010

## Addition & Subtraction.

- ▶ We can add two matrices iff they are of the same size. The  $(i, j)$ th entry of the resulting matrix is the sum of the  $(i, j)$ th entries of the two matrices being added.

### Example

$$\begin{pmatrix} 3 & -1 & 0 \\ 6 & 2 & -5 \\ 0 & -4 & 8 \end{pmatrix} + \begin{pmatrix} 4 & -1 & 6 \\ 1 & -3 & 9 \\ -2 & 0 & 8 \end{pmatrix} = \begin{pmatrix} 7 & -2 & 6 \\ 7 & -1 & 4 \\ -2 & -4 & 16 \end{pmatrix}$$

- ▶ Subtraction is defined similarly.

## Scalar Multiplication.

- ▶ We can multiply a matrix by an ordinary number, which we call a **scalar**. The product of a matrix  $A$  and a scalar  $r$ , denoted  $rA$ , is the matrix created by multiplying every entry of  $A$  by  $r$ .

### Example

$$2 \begin{pmatrix} 3 & -1 & 0 \\ 6 & 2 & -5 \\ 0 & -4 & 8 \end{pmatrix} = \begin{pmatrix} 6 & -2 & 0 \\ 12 & 4 & -10 \\ 0 & -8 & 16 \end{pmatrix}$$



## Matrix Multiplication.

- ▶ We can also multiply some matrices by other matrices.
- ▶ The product  $AB$  of two matrices  $A$  and  $B$  is only defined if the number of columns of  $A$  is equal to the number of rows of  $B$ .
- ▶ To obtain the  $(i, j)$ th entry of  $AB$ , we multiply the  $i$ th row of  $A$  and the  $j$ th column of  $B$  as shown:

$$\begin{pmatrix} a_{i1} & a_{i2} & \cdots & a_{im} \end{pmatrix} \begin{pmatrix} b_{1j} \\ b_{2j} \\ \vdots \\ b_{mj} \end{pmatrix} = a_{i1}b_{1j} + a_{i2}b_{2j} + \cdots + a_{im}b_{mj}$$

$$= \sum_h^m a_{ih}b_{hj}.$$

- ▶ The product of a  $k \times m$  and an  $m \times n$  matrix is a  $k \times n$  matrix.

# Matrix Multiplication.

## Example

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix} \begin{pmatrix} r & s & t \\ u & v & w \end{pmatrix} = \begin{pmatrix} ar + bu & as + bv & at + bw \\ cr + du & cs + dv & ct + dw \end{pmatrix}$$



- ▶ Note that we **cannot** multiply these matrices in the reverse order. **Order matters for matrix multiplication!**
- ▶ The special matrix

$$I = \begin{pmatrix} 1 & 0 & \cdots & 0 \\ 0 & 1 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & 1 \end{pmatrix},$$

has the property that  $AI = A$  for any  $m \times n$  matrix  $A$  and  $IB = B$  for any  $n \times k$  matrix  $B$ . It is called the  $n \times n$  **identity matrix**.

## Laws of Matrix Algebra.

For any matrices  $A$ ,  $B$  and  $C$ , for which the operations are defined, the following properties hold.

- ▶ Associativity:

$$\begin{aligned}(A + B) + C &= A + (B + C) \\ (AB)C &= A(BC)\end{aligned}$$

- ▶ Commutativity of Addition:

$$A + B = B + A$$

- ▶ Distributivity:

$$\begin{aligned}A(B + C) &= AB + AC \\ (A + B)C &= AC + BC\end{aligned}$$

## Laws of Matrix Algebra.

- ▶ Matrix multiplication is **not** commutative, i.e.  $AB = BA$  does not hold in general, even when both products are defined. For example

$$\begin{pmatrix} 0 & -1 \\ 1 & -1 \end{pmatrix} \begin{pmatrix} 1 & -1 \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} 0 & -1 \\ 1 & -2 \end{pmatrix},$$

but

$$\begin{pmatrix} 1 & -1 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 0 & -1 \\ 1 & -1 \end{pmatrix} = \begin{pmatrix} -1 & 0 \\ 1 & -1 \end{pmatrix}.$$

# The Transpose.

- ▶ Another frequently used operation on matrices is the **transpose**. The  $n \times m$  transpose of the  $m \times n$  matrix  $A$  is obtained by interchanging the rows and columns of  $A$  and is denoted by  $A^T$  or  $A'$ :

$$\begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{pmatrix}^T = \begin{pmatrix} a_{11} & a_{21} & \cdots & a_{m1} \\ a_{12} & a_{22} & \cdots & a_{m2} \\ \vdots & \vdots & \ddots & \vdots \\ a_{1n} & a_{2n} & \cdots & a_{mn} \end{pmatrix}.$$

- ▶ We are ‘flipping’ the matrix about its diagonal.



# The Transpose.

- ▶ Some properties of the transpose:

$$(A + B)^T = A^T + B^T$$

$$(A - B)^T = A^T - B^T$$

$$(A^T)^T = A$$

$$(rA)^T = rA^T$$

$$(AB)^T = B^T A^T$$

where  $A$  and  $B$  are of the same size and  $r$  is a scalar. Try proving the last one yourself using the definitions of the transpose and matrix multiplication (no peeking at the textbook!).

- ▶ See section 8.2 of S&B for a list of some special matrices.

# Systems of Equations in Matrix Form.

- ▶ We can now write a system of linear equations in a simplified way using matrices, as

$$Ax = b$$

where

$$A = \begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{pmatrix}, \quad x = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix}, \quad b = \begin{pmatrix} b_1 \\ b_2 \\ \vdots \\ b_m \end{pmatrix}.$$

# Algebra of Square Matrices.

## Definition.

The  $n \times n$  matrix  $B$  is an **inverse** of the  $n \times n$  matrix  $A$  if  $AB = BA = I$ .  
If the matrix  $B$  exists, we say that  $A$  is **invertible**. ▲

## Theorem (S&B 8.5)

*An  $n \times n$  matrix  $A$  can have at most one inverse.*

## Proof.

Suppose that  $B$  and  $C$  are both inverses of  $A$ . Then

$$C = CI = C(AB) = (CA)B = IB = B. \quad \blacksquare$$

- ▶ We denote this unique inverse matrix by  $A^{-1}$ .

## Algebra of Square Matrices.

### Theorem (S&B 8.6)

*If an  $n \times n$  matrix  $A$  is invertible, then it is nonsingular, and the unique solution to the system of linear equations  $Ax = b$  is  $x = A^{-1}b$ .*

### Proof.

Multiply each side of the system  $Ax = b$  by the inverse  $A^{-1}$ . Then use laws of matrix multiplication and the definition of the inverse:

$$A^{-1}(Ax) = A^{-1}b \Leftrightarrow (A^{-1}A)x = A^{-1}b \Leftrightarrow Ix = A^{-1}b \Leftrightarrow x = A^{-1}b. \quad \blacksquare$$

### Theorem (S&B 8.7)

*If an  $n \times n$  matrix  $A$  is nonsingular, then it is invertible.*

- ▶ Together, these results tell us that a square matrix is invertible iff it is nonsingular.

## Algebra of Square Matrices.

- ▶ The following theorem summarizes some facts about the inverse and the powers of matrices. By definition,

$$A^2 = AA, A^3 = AA^2, \dots A^{n+1} = AA^n.$$

### Theorem (S&B 8.10 and 8.11)

Let  $A$  and  $B$  be invertible matrices.

- (i)  $(A^{-1})^{-1} = A$ .
- (ii)  $(A^T)^{-1} = (A^{-1})^T$ .
- (iii)  $AB$  is invertible and  $(AB)^{-1} = B^{-1}A^{-1}$ .
- (iv)  $A^m$  is invertible for any  $m \in \mathbb{Z}$  and  $(A^m)^{-1} = (A^{-1})^m = A^{-m}$ .
- (v) For any  $r, s \in \mathbb{Z}$  we have  $A^r A^s = A^{r+s}$ .
- (vi) For any scalar  $r \in \mathbb{R}$  we have  $rA$  is invertible and  $(rA)^{-1} = (1/r)A^{-1}$ .

## Algebra of Square Matrices.

- ▶ One method for finding the inverse (see the proof of S&B theorem 8.7) uses elementary row operations on the augmented matrix  $[A|I]$  to reduce it to  $[I|A^{-1}]$ .

### Example

Consider any  $2 \times 2$  matrix:

$$A = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$$

- ▶ Write down the augmented matrix  $[A|I]$ :

$$[A|I] = \left( \begin{array}{cc|cc} a & b & 1 & 0 \\ c & d & 0 & 1 \end{array} \right)$$

## Algebra of Square Matrices.

- ▶ If  $a = c = 0$ ,  $A$  is singular. So suppose  $a \neq 0$ . Then we can add  $-c/a$  times row 1 to row 2, to obtain the row echelon form:

$$\left( \begin{array}{cc|cc} a & b & 1 & 0 \\ 0 & \frac{ad-bc}{a} & -\frac{c}{a} & 1 \end{array} \right)$$

- ▶ This shows that  $A$  is nonsingular iff  $ad - bc \neq 0$ .
- ▶ Now multiply row 1 by  $1/a$  and row 2 by  $a/(ad - bc)$  to get

$$\left( \begin{array}{cc|cc} 1 & \frac{b}{a} & \frac{1}{a} & 0 \\ 0 & 1 & -\frac{\frac{c}{a}}{ad-bc} & \frac{1}{ad-bc} \end{array} \right).$$

## Algebra of Square Matrices.

- ▶ After adding  $-b/a$  times row 2 to row, we obtain the reduced row echelon matrix:

$$\left( \begin{array}{cc|cc} 1 & 0 & \frac{d}{ad-bc} & -\frac{b}{ad-bc} \\ 0 & 1 & -\frac{c}{ad-bc} & \frac{a}{ad-bc} \end{array} \right).$$

- ▶ So the inverse of  $A$  is:

$$A^{-1} = \frac{1}{ad-bc} \begin{pmatrix} d & -b \\ -c & a \end{pmatrix}.$$

- ▶ Note that if  $ad - bc \neq 0$ ,  $a$  and  $c$  cannot both be zero. If  $a = 0$ , we must have  $c \neq 0$  and we can go through steps similar to above to obtain an inverse. ◆



# The Determinant of a Matrix.

- ▶ The **determinant** of a matrix  $A$  is denoted  $|A|$  or  $\det(A)$ .
- ▶ The determinant is a number we can compute for any square matrix to determine whether a matrix is nonsingular (i.e. the corresponding linear system has exactly one solution for every choice of RHS  $b$ ).
- ▶ A  $1 \times 1$  matrix (i.e. a scalar)  $a$  has an inverse  $1/a$  iff  $a \neq 0$ . So we define the determinant of  $a$  as:

$$\det(a) = a$$

# The Determinant of a Matrix.

- ▶ For a  $2 \times 2$  matrix (see S&B example 8.4):

$$\begin{aligned} \begin{vmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{vmatrix} &= a_{11}a_{22} - a_{12}a_{21} \\ &= a_{11} \det(a_{22}) - a_{12} \det(a_{21}). \end{aligned}$$

## Example

$$\begin{vmatrix} 5 & -3 \\ 2 & -1 \end{vmatrix} = 5(-1) - (-3)2 = 1$$



# The Determinant of a Matrix.

## Definition.

Let  $A_{ij}$  be the  $(n - 1) \times (n - 1)$  submatrix formed by deleting row  $i$  and column  $j$  from the  $n \times n$  matrix  $A$ . Then the scalar

$$M_{ij} = \det A_{ij}$$

is called the  $(i, j)$ th **minor** of  $A$ , and the scalar

$$C_{ij} = (-1)^{i+j} M_{ij}$$

is called the  $(i, j)$ th **cofactor** of  $A$ . ▲

- ▶ Remember the signs of the cofactors by arraying them in a matrix:

$$\begin{pmatrix} + & - & + & \cdots \\ - & + & - & \cdots \\ + & - & + & \cdots \\ \vdots & \vdots & \vdots & \ddots \end{pmatrix}.$$

## The Determinant of a Matrix.

- Now we can write the determinant of a  $2 \times 2$  matrix as

$$\begin{aligned} \begin{vmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{vmatrix} &= a_{11} \det(a_{22}) - a_{12} \det(a_{21}) = a_{11}C_{11} + a_{12}C_{12} \\ &= a_{11}M_{11} - a_{12}M_{12}. \end{aligned}$$

- The determinant of a  $3 \times 3$  matrix  $A$  is given by:

$$\begin{aligned} \begin{vmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{vmatrix} &= a_{11}C_{11} + a_{12}C_{12} + a_{13}C_{13} \\ &= a_{11}M_{11} - a_{12}M_{12} + a_{13}M_{13} \\ &= a_{11} \begin{vmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{vmatrix} - a_{12} \begin{vmatrix} a_{21} & a_{23} \\ a_{31} & a_{33} \end{vmatrix} \\ &\quad + a_{13} \begin{vmatrix} a_{21} & a_{22} \\ a_{31} & a_{32} \end{vmatrix} \end{aligned}$$

# The Determinant of a Matrix.

## Example

$$\begin{aligned}
 \begin{vmatrix} 1 & -2 & 3 \\ -3 & 4 & 2 \\ -1 & -3 & 6 \end{vmatrix} &= a_{11}C_{11} + a_{12}C_{12} + a_{13}C_{13} \\
 &= 1(-1)^{(1+1)} \begin{vmatrix} 4 & 2 \\ -3 & 6 \end{vmatrix} + (-2)(-1)^{(1+2)} \begin{vmatrix} -3 & 2 \\ -1 & 6 \end{vmatrix} \\
 &\quad + 3(-1)^{(1+3)} \begin{vmatrix} -3 & 4 \\ -1 & -3 \end{vmatrix} \\
 &= 1(1)[4(6) - (2)(-3)] + (-2)(-1)[-3(6) - 2(-1)] \\
 &\quad + (1)3[(-3)(-3) - 4(-1)] \\
 &= 1(30) + (-2)(16) + 3(13) \\
 &= 37
 \end{aligned}$$

# The Determinant of a Matrix.

## Definition.

The **determinant** of an  $n \times n$  matrix  $A$  is given by:

$$\begin{aligned}\det A &= a_{11}C_{11} + a_{12}C_{12} + \cdots + a_{1n}C_{1n} \\ &= a_{11}M_{11} - a_{12}M_{12} + \cdots + (-1)^{n+1}a_{1n}M_{1n}.\end{aligned}$$



- ▶ In fact, it turns out that we can expand along any row or column of the matrix to calculate the determinant.
- ▶ For example, expanding along the  $j$ th column:

$$\begin{aligned}\det A &= a_{1j}C_{1j} + a_{2j}C_{2j} + \cdots + a_{nj}C_{nj} \\ &= (-1)^{j+1}[a_{1j}M_{1j} - a_{2j}M_{2j} + \cdots + (-1)^{n+1}a_{1n}M_{1n}].\end{aligned}$$

This can come in useful if it is easier to expand along one row or column because, for instance, it has a lot of zeros.

## The Determinant of a Matrix.

- ▶ Some useful facts can simplify the calculation of the determinant.

### Theorem (S&B 9.1,9.2 and 9.5)

*Let  $A$  and  $B$  be square matrices, then*

- (i)  $|A^T| = |A|$ .
- (ii)  $|AB| = |A||B|$ .
- (iii)  $|A + B| \neq |A| + |B|$ .
- (iv) *If  $B$  comes from interchanging two rows or columns of  $A$ , then  $|B| = -|A|$ .*
- (v) *If  $A$  any row (column) is a multiple of another row (column), then  $|A| = 0$ .*
- (vi) *If  $B$  comes from adding a multiple of one row to another row, then  $|C| = |A|$ .*
- (vii) *If  $|A|$  is a lower-triangular, upper-triangular, or diagonal matrix, then its determinant is simply the product of its diagonal entries.*

## The Determinant of a Matrix.

- ▶ We now put together some facts about square matrices.

### Theorem (S&B 8.9 and 9.3)

*For any  $n \times n$  matrix  $A$ , the following statements are equivalent:*

- (1)  *$A$  is invertible.*
- (2)  *$A$  has a right inverse.*
- (3)  *$A$  has a left inverse.*
- (4) *Every system  $Ax = b$  has at least one solution for every  $b$ .*
- (5) *Every system  $Ax = b$  has at most one solution for every  $b$ .*
- (6)  *$A$  is nonsingular.*
- (7)  *$A$  has maximal rank  $n$ .*
- (8)  *$A$  has a nonzero determinant.*



## Uses of the Determinant.

### Definition.

The **adjoint** of a matrix  $A$  is the transpose of the matrix of cofactors of  $A$ :

$$\begin{pmatrix} C_{11} & C_{21} & \cdots & C_{n1} \\ C_{12} & C_{22} & \cdots & C_{n2} \\ \vdots & \vdots & \ddots & \vdots \\ C_{1n} & C_{2n} & \cdots & C_{nn} \end{pmatrix}$$



### Theorem

Let  $A$  be an  $n \times n$  nonsingular matrix. Then:

(i)

$$A^{-1} = \frac{\text{adj } A}{|A|}, \quad \text{and}$$

## Uses of the Determinant.

- (ii) (**Cramer's rule**) the unique solution  $x = (x_1, \dots, x_n)$  of the system  $Ax = b$  is

$$x_i = \frac{|B_i|}{|A|}, \quad \text{for } i = 1, \dots, n,$$

where  $B_i$  is the matrix  $A$  with the RHS  $b$  replacing the  $i$ th column of  $A$ .

### Example

We will calculate the inverse of the matrix  $A$  (from our previous example), where

$$A = \begin{pmatrix} 1 & -2 & 3 \\ -3 & 4 & 2 \\ -1 & -3 & 6 \end{pmatrix}.$$

## Uses of the Determinant.

- ▶ Since we already calculated  $|A|$ , all we need to do is calculate  $\text{adj } A$
- ▶ We calculated  $C_{11} = 30$ ,  $C_{12} = 16$  and  $C_{13} = 13$  to find  $|A|$ , we now find the other cofactors.

$$C_{21} = (-1)^{2+1} \begin{vmatrix} -2 & 3 \\ -3 & 6 \end{vmatrix} \qquad C_{22} = (-1)^{2+2} \begin{vmatrix} 1 & 3 \\ -1 & 6 \end{vmatrix} \\ = 3 \qquad \qquad \qquad = 9$$

$$C_{23} = (-1)^{2+3} \begin{vmatrix} 1 & -2 \\ -1 & -3 \end{vmatrix} \\ = 5$$

## Uses of the Determinant.

$$C_{31} = (-1)^{3+1} \begin{vmatrix} -2 & 3 \\ 4 & 2 \end{vmatrix} \quad C_{32} = (-1)^{3+2} \begin{vmatrix} 1 & 3 \\ -3 & 2 \end{vmatrix}$$

$$= -16 \qquad \qquad \qquad = -11$$

$$C_{33} = (-1)^{3+3} \begin{vmatrix} 1 & -2 \\ -3 & 4 \end{vmatrix}$$

$$= -2$$

- Now arrange these in the matrix of cofactors

$$\begin{pmatrix} 30 & 16 & 13 \\ 3 & 9 & 5 \\ -16 & -11 & -2 \end{pmatrix}$$

## Uses of the Determinant.

- ▶ And take the transpose to get the adjoint

$$\text{adj } A = \begin{pmatrix} 30 & 3 & -16 \\ 16 & 9 & -11 \\ 13 & 5 & -2 \end{pmatrix}$$

- ▶ Then the inverse is

$$A^{-1} = \frac{\text{adj } A}{|A|} = \frac{1}{37} \begin{pmatrix} 30 & 3 & -16 \\ 16 & 9 & -11 \\ 13 & 5 & -2 \end{pmatrix}$$

- ▶ Finally, check that  $AA^{-1} = I$ .



# Uses of the Determinant.

## Example

For  $3 \times 3$  systems (also, see example 9.4 S&B)

$$a_{11}x_1 + a_{12}x_2 + a_{13}x_3 = b_1$$

$$a_{21}x_1 + a_{22}x_2 + a_{23}x_3 = b_2$$

$$a_{31}x_1 + a_{32}x_2 + a_{33}x_3 = b_3,$$

Cramer's rule states that

$$x_1 = \frac{\begin{vmatrix} b_1 & a_{12} & a_{13} \\ b_2 & a_{22} & a_{23} \\ b_3 & a_{32} & a_{33} \end{vmatrix}}{\begin{vmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{vmatrix}}, \quad x_2 = \frac{\begin{vmatrix} a_{11} & b_1 & a_{13} \\ a_{21} & b_2 & a_{23} \\ a_{31} & b_3 & a_{33} \end{vmatrix}}{\begin{vmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{vmatrix}}, \quad x_3 = \frac{\begin{vmatrix} a_{11} & a_{12} & b_1 \\ a_{21} & a_{22} & b_2 \\ a_{31} & a_{32} & b_3 \end{vmatrix}}{\begin{vmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{vmatrix}}$$